

# Augenblick, Lazarus: Restrictions on Asset-Price Movements Under Rational Expectations: Theory and Evidence

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1. Beliefs, risk adjustments and asset prices
  - information in no-arbitrage restrictions
2. Summary of the paper methodology and results
  - martingale restrictions on beliefs
  - localization of state prices
  - implied slopes of the SDF
3. Role of assumptions
  - key assumption: constant SDF slope
4. Questions and suggestions
  - are we testing for rational expectations?

High volatility in prices of risky securities

- high frequency (intraday, daily returns)
- business cycle frequency (monthly, quarterly returns)

Are these asset price fluctuations 'rational'?

- i.e., can they be explained by a plausible model of investors' preferences, market constraints, and investors' beliefs that are correct (in line with the data-generating process (DGP))?

Abstract from market constraints, assume no arbitrage.

- then, for any investor's subjective belief  $\hat{P}$ , there exists an SDF  $m_{t+1}$  that prices financial assets

$$1 = \hat{E}_t [m_{t+1} R_{t+1}]. \quad (1)$$

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Which is the right model of the SDF,  $m_{t+1}$  or  $\hat{m}_{t+1}$ ?

- that depends on what is plausible  $\implies$  impose structure on the SDF
- asset price data alone cannot tell (1) and (2) apart
  - even if we perfectly know the DGP measure  $P$

Information from no-arbitrage restrictions

$$1 = E_t[\underbrace{m_{t+1}h_{t+1}}_{\hat{m}_{t+1}}R_{t+1}]$$

- **Time series information** identifies the DGP measure  $P$ .
- **Data on returns**  $R_{t+1}$  together with **no-arbitrage restrictions** inform  $\hat{m}_{t+1}$ .
- We must impose **additional restrictions** to separately infer something about  $m_{t+1}$  and  $h_{t+1}$ .

This paper

$$1 = E_t[\underbrace{m_{t+1}h_{t+1}}_{\hat{m}_{t+1}}R_{t+1}]$$

- imposes the zero of **correct beliefs**  $\implies h_{t+1} \equiv 1$
- restricts the class of SDFs (**conditional transition independence (CTI)**)  
 $\implies$  restriction on  $m_{t+1}$
- uses **martingale restrictions under DGP** (time series information) to deduce whether  $\exists m_{t+1}$  in the given class that also prices returns



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## Outcomes

- if  $\exists$  such  $m_{t+1}$ , characterize the minimum needed local risk aversion consistent with  $h_{t+1} \equiv 1$
- if  $\nexists$  such  $m_{t+1}$ , conclude that  $h_{t+1} \neq 1$ .

Let  $X_t$  be a square-integrable martingale. Then

$$E_t \left[ \sum_{j=t}^{T-1} (X_{j+1} - X_j)^2 \right] = E_t [X_T^2 - X_t^2]$$

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# MARTINGALE RESTRICTIONS

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‘Beliefs are martingales’ (under their own measure)

- apply the above result to a two-state outcome  $\implies$  localization
- $\pi_t$  – probability of the low state realization at  $T$

$$E_t [m_T] \doteq E_t \left[ \sum_{j=t}^{T-1} (\pi_{j+1} - \pi_j)^2 \right] = E_t [(1 - \pi_t) \pi_t - (1 - \pi_T) \pi_T] \doteq E_t [r_T]$$

- if we could measure beliefs, this would constitute a test of rational expectations (Augenblick and Rabin (2018))
- restriction on how much beliefs can vary over time

Instead, we observe risk-neutral (Arrow-Debreu) prices  $\pi_t^*$ . Then

$$E_t[m_T^*] \doteq E_t \left[ \sum_{j=t}^{T-1} (\pi_{j+1}^* - \pi_j^*)^2 \right] \neq E_t [(1 - \pi_t^*) \pi_t^* - (1 - \pi_T^*) \pi_T^*] \doteq E_t[r_T^*]$$

Relationship between  $\pi_t^*$  and  $\pi_t$

$$\pi_t^* \propto E_t[M_T/M_t \mid s_T = \text{low}] \pi_t$$

- reflects risk adjustment and normalization by the risk-free rate

Rewrite relation between  $\pi_t^*$  and  $\pi_t$  in terms of odds ratio

$$\underbrace{\frac{\pi_t^*}{1 - \pi_t^*}}_{\text{risk-neutral prices}} = \underbrace{\frac{E_t [M_T/M_t \mid s_T = \text{low}]}{E_t [M_T/M_t \mid s_T = \text{high}]}}_{\text{SDF slope } \phi} \underbrace{\frac{\pi_t}{1 - \pi_t}}_{\text{beliefs}}$$

If we observe (a lot of) variation in  $\pi_t^*$  over time, what can we infer about variation in  $\pi_t$ ?

- we must restrict movements in the SDF slope  $\phi$  over time
- $\phi$  can differ across terminal state realizations

Assumption: **Conditional transition independence (CTI)**

- $\phi$  is constant over time (it can also follow a martingale)

$$\underbrace{\frac{\pi_t^*}{1 - \pi_t^*}}_{\text{risk-neutral prices}} = \underbrace{\frac{E_t [M_T/M_t \mid s_T = \text{low}]}{E_t [M_T/M_t \mid s_T = \text{high}]}}_{\text{SDF slope } \phi} \underbrace{\frac{\pi_t}{1 - \pi_t}}_{\text{beliefs}}$$

**Idea:** A high  $\phi$  translates large variation in  $\pi_t^*$  into modest variation in  $\pi_t$ , which can then satisfy the martingale restriction on correct beliefs.

$$E_t [m_T^* - r_t^*] \leq (\pi_t^*)^2 \left( 1 - \frac{1}{\pi_t^* + \phi(1 - \pi_t^*)} \right)$$

Define states: Terminal values of a market index  $R_T^m = s_T$  at some date  $T$ .

Extract Arrow–Debreu prices (Breedon, Litzenberger (1978))

Construct sample averages for  $E_t[m_T^* - r_t^*]$

- daily, weekly, monthly frequencies
- conditional on terminal returns for various adjacent payoff states  $R_T^m \in \{s_j, s_{j+1}\} \implies$  localization of the return distribution

Infer the required  $\phi$ 's to satisfy inequality

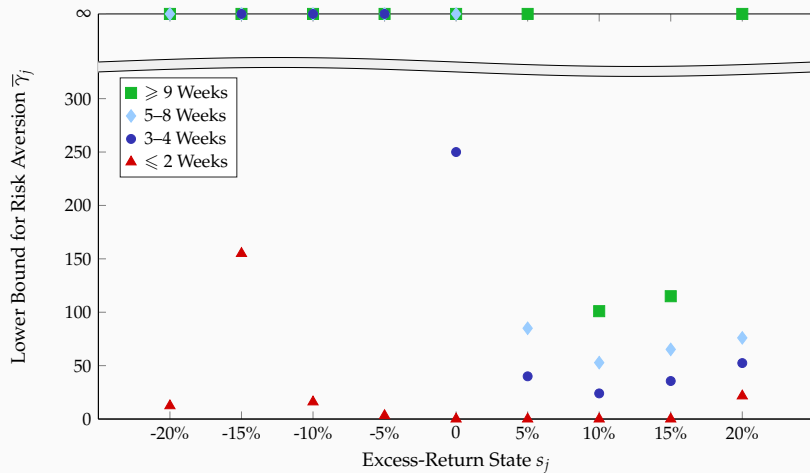
$$E_t[m_T^* - r_t^*] \leq (\pi_t^*)^2 \left( 1 - \frac{1}{\pi_t^* + \phi(1 - \pi_t^*)} \right)$$

Required SDF slopes  $\phi$  need to be very high to rationalize correct beliefs.

- Consequence of high volatility in  $\pi_t^*$ .



Figure 5: Estimates of Relative Risk Aversion: Splits by Time to Expiration



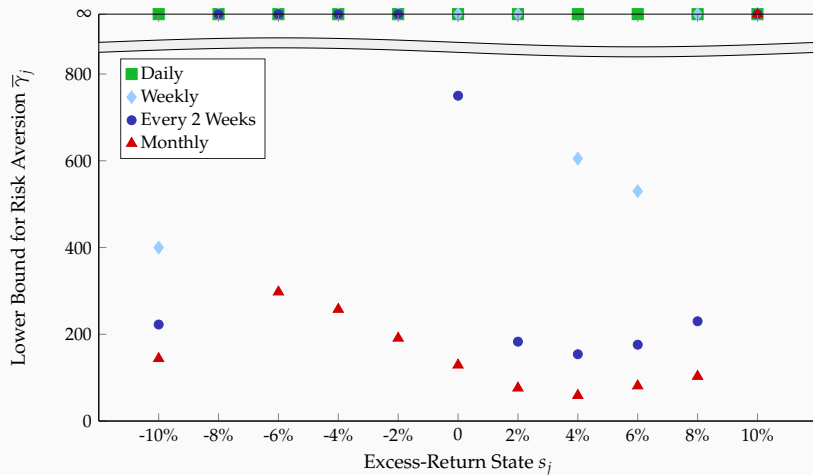
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Higher  $\phi$  needed to rationalize higher frequency (daily) variation.

- Is there more 'unexplained' variation at higher frequencies? Unclear, only a necessary bound.

Figure 6: Estimates of Relative Risk Aversion: Splits by Sampling Frequency



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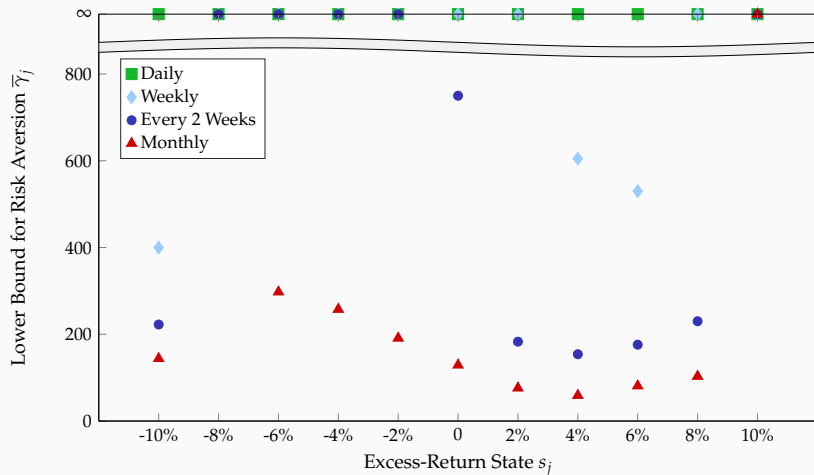
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High risk aversion needed across all terminal payoff states.

- Informative about the type of risks faced by investors (e.g., tail risk).

Figure 6: Estimates of Relative Risk Aversion: Splits by Sampling Frequency



## DISCUSSION OF ASSUMPTIONS

The whole identification hinges on the choice of class of 'plausible' SDFs.

$$\frac{\pi_t^*}{1 - \pi_t^*} = \frac{E_t [M_T / M_t \mid R_T^m = \text{low}]}{\underbrace{E_t [M_T / M_t \mid R_T^m = \text{high}]}_{\text{constant SDF slope } \phi}} \frac{\pi_t}{1 - \pi_t}$$
$$\phi = \frac{u'(R_T^m = \text{low})}{u'(R_T^m = \text{high})}$$

To first-order approximation, this corresponds to **constant local relative risk aversion**  $\gamma(\bar{R})$

- market risk premium

$$E_t [R_T^e] = \gamma(\bar{R}) \text{Var}_t [R_T^e]$$

- $\gamma(\bar{R})$  is the local curvature of the utility/value function as a function of the return realization, by assumption independent of  $t$
- time-variation in market risk premium can only be driven by  $\text{Var}_t [R_T^e]$

## WHEN DOES THE ASSUMPTION HOLD?

### Separable utility over returns at terminal state

- **caveat**: if utility is over consumption, then time variation in the mapping between returns and consumption can lead to violations

### Epstein–Zin in iid growth environment

- not very interesting, isomorphic to recalibrated separable CRRA utility

### Disaster risk model (Gabaix (2012)) with CRRA utility

- holds for (high) return realizations conditional on which disasters between today and maturity have a negligible chance
- **interesting**: utilizes localization to return dynamics of particular AD securities
- for a rejection of rationality, it requires establishing what negligible chance means in the model and data

In general,  $\phi$  cannot move systematically with  $\pi_t^*$ .

$$\frac{\pi_t^*}{1 - \pi_t^*} = \underbrace{\frac{E_t [M_T/M_t \mid R_T^m = \text{low}]}{E_t [M_T/M_t \mid R_T^m = \text{high}]}}_{\text{constant SDF slope } \phi} \frac{\pi_t}{1 - \pi_t}$$

## Inference with constant $\phi$

- bad times  $\implies$  observed  $\uparrow \pi_t^* \implies$  infer required increase in  $\pi_t$ , given  $\phi$

## Systematic variation in $\phi$

- bad times  $\implies$  observed  $\uparrow \pi_t^* \implies$  associated increase in  $\phi \implies$  lower required increase in  $\pi_t$
- lower average  $\phi$  needed to dampen implied variation in  $\pi_t$  sufficiently to satisfy the martingale restriction



## WHAT IS OFF THE TABLE?

**Nonseparable preferences** (Epstein–Zin) with interesting state dynamics

Explicit **time-variation in risk aversion**

Models with agent heterogeneity (e.g., in risk aversion)

- implied risk aversion a function of wealth distribution
- **how to interpret results in models with heterogeneous beliefs?**

Models with **financial constraints**

- implied risk aversion changes with tightness of constraints

**Habit formation models** (Campbell and Cochrane (1999))

- implied risk aversion function of habit level
- authors show that in CC (1999), the bound still holds approximately
  - but SDF slopes implied by the model are trivial for non-extreme states

$$\frac{\pi_t^*}{1 - \pi_t^*} = \frac{E_t [M_T/M_t \mid R_T^m = \text{low}]}{\underbrace{E_t [M_T/M_t \mid R_T^m = \text{high}]}_{\text{time-varying SDF slope } \phi_t}} \frac{\pi_t}{1 - \pi_t}$$

**Question:** How much variation in  $\phi_t$  is needed such that implied  $\pi_t$  satisfy the martingale restriction?

- a test of a larger class of (**more interesting**) models

Existing models may well still fail

- many asset pricing models designed to match unconditional risk premia and the cross-section
- appropriate time-variation harder to achieve

## OTHER CONSIDERATIONS

Implementation relies purely on **option prices**

- studies option price dynamics (with all their specifics) relative to the index
- what about correct beliefs in the pricing of the index itself?

Authors relate identified excess belief movement to macro statistics

- why not **survey data** directly?

Why bounds at all?

- Given the information obtained from localization, one could directly attempt to **estimate the SDF nonparametrically** (Christensen (2017))

Approximation details (discretization, etc.)

- authors take these seriously, but details are subtle

Localization

- we do not need to test adjacent states only, additional information in alternative state combinations

Given stated assumptions, the paper is cleanly executed, with a lot of detail.

- robustness checks, careful econometrics
- models could be studied in more detail

Interesting results

- much higher **high frequency** (daily) **variation** than variation at 'macro' frequencies
  - Market microstructure effects in daily data? Different models needed?
  - Caution regarding bid-ask bounce? (not covered here)
- **localization of return dynamics** can discard some models
  - high AD price variation for high-return states speaks against disaster-type models that focus risk adjustments to adverse states

### A test of rational expectations?

- restricts attention to a class of models that is too narrow
- reverting the problem and studying required variation in implied risk aversion may be more fruitful

Even if we impose stringent restrictions on SDF and reject correct beliefs, **the puzzle still stands.**

- **why do beliefs fluctuate so much?**
- need to devise interesting **models of belief dynamics** to make progress ...
- ... just like interesting **models of risk adjustments** that go beyond imposed restrictions ...
- ... and in some cases, the two **cannot be distinguished** using asset price data anyway
  - e.g. Epstein–Zin with unitary IES vs. Hansen–Sargent robust preference vs. ex-post Bayesian worst-case belief

## SIDE NOTE: RELATIONSHIP TO ROSS (2015)

Ross (2015) 'Recovery Theorem'

- uses only cross-sectional price information, no time series
- imposes separable marginal utility (transition independence (TI))

$$m_{t+1} = \beta \frac{u'(s_{t+1})}{u'(s_t)}$$

and stationarity of the state dynamics

- under TI, it recovers subjective belief  $\hat{P}$  (possibly different from  $P$ )
- if TI does not hold, it recovers a long-run risk neutral measure (different from  $\hat{P}$  or  $P$ )

This paper

- imposes a restriction based on time-series information in  $P$  (time-variation in prices)
- asks whether imposing  $P = \hat{P}$  can be consistent with SDFs that satisfy CTI
- allows for nonstationarity of the state and hence for some martingale components in SDF
- both papers assume time-invariant local curvature over terminal states